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Elementary transformation analysis for Array-OL

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Abstract: Array-OL is a high-level specification language dedicated to the definition of intensive signal processing applications. Several tools exist for implementing an Array-OL specification as a data parallel program. While Array-OL can be used directly, it is often convenient to be able to deduce part of the specification from a sequential version of the application. This paper proposes such an analysis and examines its feasibility and its limits.

Key-words: Array-OL, multidimensional signal processing, program analysis

Analyse de transformations élémentaires pour Array-OL

Résumé : Array-OL est un système de spécification de haut niveau spécialisé dans la définition d'application de traitement du signal intensif. Il existe plusieurs ateliers qui transforment une spécification Array-OL en un programme à parallélisme de données. Bien que Array-OL puisse être utlisé tel quel, il est souvent intéressant de pouvoir déduire ses paramètres d'une version séquentielle de l'application. Ce rapport propose une telle analyse et en examine la faisabilité et les limites.

Mots-clés : Array-OL, traitement du signal multidimensionnel, analyse de programme

1   Introduction

In the Array-OL formalism [1, 2], a program is a network of processes which communicate through shared arrays. A process is made of one or more parallel loops. At each iteration of these loops, a task (or elementary transform) is executed. The elementary transform may contain one or more loops, which are executed sequentially.

The execution of an elementary task can be decomposed into three steps:

- Move portions of the input array(s) (regions) to the local memory of the processor executing the task.

- Execute the elementary transform and generate portions of the output array(s).

- Move the results to the output array(s).

In order to simplify code generation, the input and output regions must move uniformly across the shared arrays. It is admissible that each elementary transform use only a subset of regularly spaced entries in the input and output regions. In the present version of the software, regions must not overlap, as this would precludes parallel execution of the outer loops. The useful elements of a region are collected in a pattern, which must be a rectangular parallelepiped of fixed size.

The Array-OL formalism may be used directly. The programmer is responsible for constructing the elementary transform, identifying the input and output regions, checking parallelism and specifying the regions parameters. Another possibility is to infer the Array-OL specification from a sequential version of the program. This requires the solution of three problems:

- Rewriting the sequential program in such a way that the outer loops have no dependences.

- Deducing the shape and size of the regions from an analysis of the array subscript functions.

- Rewriting the sequential code by substituting pattern accesses to the original array accesses.

This note is dedicated to a proposal for the solution of the second and third problems. The assumption is that one is given the sequential code, together with a list of input and output arrays, and an indication of which loop(s) are to be considered as the outer (repetition) loop(s).

2   Paving

Let A be an input or output array and let its occurences in the sequential code be numbered from 1 to N. Let r be the counter(s) of the repetition loop(s), and let j^k be the counter(s) of the inner loop(s) that surround occurence k of A. Let $e^k(r, j^k)$ be its subscript function. e^k is a vector function whose dimension is the rank of A.

To be amenable to an Array-OL implementation, the subscript function e^k must be affine in r and j^k. A convenient way of checking this property consists in computing the two Jacobian matrices:

$$P^k = (\frac{\partial e^k_\alpha}{\partial r_\beta}) \quad B^k = (\frac{\partial e^k_\alpha}{\partial j^k_\beta}),$$

checking that they do not depend on r or j^k, and verifying the identity:

$$e^k(r, j^k) = P^k r + B^k j^k + e^k(0, 0).$$

In Array-OL terminology, P^k is the paving matrix, and $e^k(0, 0)$ is the origin of the paving. The elements of these entities may be numbers, or they may depend on constants, which must be given numerical values just before code generation. References with different paving matrices may be separated by arbitrary distance in the source or target array; it is not possible to group them efficiently; they must be implemented as separate channels.

In the following example:

```
myTE( in[][], out[]){
for(i=0;i<7; i++)  // boucle TE
{
for ( k=0;k<11;k++)
{
S=0;
    for(j=0;j<100;j++)
    {
    S+=  in[0][j+11] * in[i+1][k+j];
    }
    out[ i][k]=S;
}
}
```
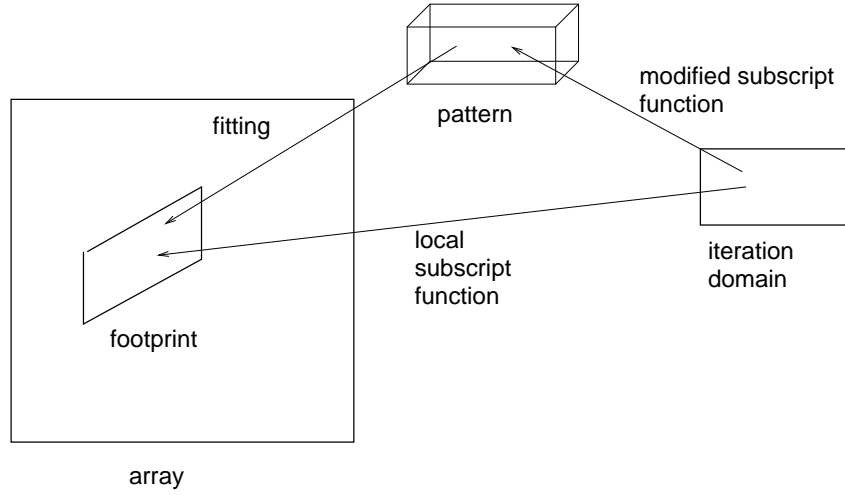
Figure 1: Data access in Array-OL

there are two references to `in` with repective subscript functions $e^1(i,k,j) = \begin{pmatrix} 0 \\ j+11 \end{pmatrix}$ and $e^2(i,k,j) = \begin{pmatrix} i+1 \\ k+j \end{pmatrix}$. The corresponding paving matrices are $P^1 = \begin{pmatrix} 0 \\ 0 \end{pmatrix}$ and $P^2 = \begin{pmatrix} 1 \\ 0 \end{pmatrix}$. Hence, the two accesses must be handled separately.

In the following, I assume that accesses to A have been partitioned according to their paving matrix, and consider only one partition at a time. The size of the repetition space is deduced simply from the bound(s) of the elementary transform loop(s). In the Spear/DE implementation of Array-OL, there may be further constraints on the paving matrix (e.g. that it be a permutation of a diagonal matrix).

3 Pattern and fitting

A pattern is a compact specification of all the elements of an array that are accessed, with references having the same paving matrix, in one iteration of the external loop(s).

When discussiong patterns, one has to consider three frames of reference (see Fig. 1). The first one is the original (input or output) array. Its dimension is the

rank of the array, noted $|A|$, and its coordinates are called *subscripts*. The shape of an array is always a (hyper-) rectangle.

The second frame of reference is the iteration space of the inner loops of the elementary transform. Its dimension is the number of loops enclosing the reference, noted d^k, and its coordinates are called *loop counters*. There may be as many iteration domains as there are references, or several references may share the same iteration domain. The shape of an iteration domain is arbitrary. The only requirement in the present context is to be able to construct its vertices, either because the iteration domain is rectangular, or because it can be expressed as a convex polyhedron with parameters in the constant terms only. The iteration domain of reference k will be denoted as D^k in what follows.

The third frame of reference is the pattern. According to Boulet [1] the pattern is always of rectangular shape. The pattern associated to reference k is denoted by T^k and its dimension is p^k. The associated fitting matrix, F^k, connects the pattern space to the array space and its dimension, accordingly, is $|A| \times p^k$.

The relation of these objects are as follows. Firstly, the local subscript function $f^k(j^k) = B^k j^k + e^k(0,0) = e^k(0, j^k)$ gives the coordinates of an array cell relative to the reference point $P^k.r$ which moves according to the paving matrix.

Next, the image $f^k(D^k)$ is the *footprint* of reference k. Its shape is arbitrary. The images of the vertices of D^k by f^k form a superset of the vertices of the footprint; a representation as a convex polyhedron can be recovered by one application of the Chernikova algorithm [3].

Lastly, the image of the pattern by the fitting matrix must enclose the footprint, and it must be feasible to retrieve a datum from the pattern instead of the original array. This implies that there exists a function ϕ^k from D^k to T^k such that for every iteration vector $j^k \in D^k$, $f^k(j^k) = F^k \phi^k(j^k)$. In the text of the elementary transform, ϕ^k must be substituted to e^k in reference k to A.

As one may see from this discussion, while the iteration domain and footprint are fixed once the sequential program is given, the choice of the pattern and fitting matrix are somewhat arbitrary. There are two obvious solutions: in the first one, the pattern is the smallest rectangular box enclosing the footprint, the fitting matrix is the identity, and the subscript function is not changed. In the second solution, the pattern is isomorphic to the iteration domain (provided it is a parallelepiped), B^k is the fitting matrix, and the new subscript function is the identity.

In signal processing applications, it is often the case that several references to the same array have similar subscript functions; constructing only one pattern for several references is an interesting optimization. However, this should not be obtained at

the cost of a large overhead in the size of the pattern. In other word, the number of useless elements in the pattern must be minimized. Useless elements come from two sources:

- A subscript matrix which is not of full row rank: the pattern will have more dimensions than the footprint.

- A subscript matrix whose determinant is not of modulus one: there will be holes (unused elements) in the footprint. The inverse of the determinant gives an asymptotic evaluation of the ratio of useful elements.

The next section presents a method for computing a pattern and a fitting matrix in the general case (many references). This method can only be applied if all elements of the matrices B^k and the vectors b^k have known numerical values. Section 5 presents fail-soft solutions for cases in which these elements depend on unknown parameters.

4   The General Case

The basic observation is that a conservative estimate of the footprint can be obtained by computing the projection of each iteration domain by the associated subscript function, then constructing a convenient superset of the union of these projections. One practical method consists in projecting the vertices of the iteration domains. One then gathers all such projections, and constructs their convex hull by familiar (e.g., Chernikova's) algorithms.

To reduce the size overhead, one should notice that a useful point for reference k also belongs to the lattice which is generated by the column vectors of B^k. Hence, B^k, properly simplified (see later) could be used as the fitting matrix. However, in the case of several references, we have to combine several lattices into one, since each pattern has only one fitting matrix. As an illustration of this construction, consider the one dimensional case. A one-dimensional lattice is simply a set of regularly spaced points. Combining two lattices generates a lattice whose spacing is the gcd of the component spacings. The many-dimensional equivalent of the gcd is the construction of the Hermite normal form of the subscript matrices.

Let $\Lambda(B, b)$ be the lattice generated by B with origin b, i.e. the set of points $\{Bx + b \mid x \in \mathbb{N}^d\}$. Let $L^1 = \Lambda(B^1, b^1)$ and $L^2 = \Lambda(B^2, b^2)$ be two such lattices. I claim that the union of L^1 and L^2 is included in the lattice $L = \Lambda([B^1 B^2 (b^2 - b^1)], b^1)$.

Proof Let $B^1.x + b^1$ be a point of L^1. We have:

$$B^1.x + b^1 = B^1.x + B^2.0 + (b^2 - b^1).0 + b^1$$

hence $B^1.x + b^1$ is in L. Similarly:

$$B^2.y + b^2 = B^1.0 + B^2.y + (b^2 - b^1).1 + b^1.$$

I conjecture that L is the smallest lattice which includes L^1 and L^2. The proof is obvious if the bs are null. The general case is left for future work. ∎

The construction can be extended to any number of component lattices. The resulting matrix is $[B^1 \ldots B^N (b^2 - b^1) \ldots (b^N - b^1)]$ and the origin is b^1. Furthermore, b^1 can be moved to the origin of the paving and hence taken as 0 when computing the fitting.

In case where B has been obtained by mixing many references, it must be simplified before being used for an Array-OL specification.

The starting point of this simplification is the *row echelon* form of B. One can show (see the appendix) that there exists two unitary matrices P and U such that:

$$B = P \begin{bmatrix} H & 0 \\ C & 0 \end{bmatrix} U,$$

where H is a square upper triangular matrix of size $r \times r$ with positive diagonal coefficients, C is arbitrary, and both 0 represent null matrices of appropriate sizes. r is the row rank of B. Furthermore, U can be partitioned, row wise, in two matrices of size $r \times d$ and $(d - r) \times d$, $U = \begin{bmatrix} U' \\ U'' \end{bmatrix}$.

Let j be a point in the iteration domain of the inner loops. The corresponding point in the footprint is:

$$\begin{aligned} Bj &= P \begin{bmatrix} H & 0 \\ C & 0 \end{bmatrix} \begin{bmatrix} U' \\ U'' \end{bmatrix} j & (1) \\ &= P \begin{bmatrix} H \\ C \end{bmatrix} (U'j) & (2) \end{aligned}$$

One possible interpretation of this formula is that the pattern for the current reference is the image of its iteration domain by U', and that the corresponding paving

matrix is $P \begin{bmatrix} H \\ C \end{bmatrix}$. In the body of the elementary transform, accesses to Bj in the input or output array have to be replaced by accesses to $U'j$ in the pattern. It may be that the pattern computed in this way is not rectangular, in which case it must be "boxed" by computing the component-wise minima and maxima of its extreme points. The dimension of the pattern is r.

It is interesting to notice that this general solution reduces to one of the approximate methods above in special cases. If B is unitary, then its row echelon form is the unit matrix. In that case, the pattern is the footprint, eventually extended to a rectangular box and the fitting matrix is the identity. Conversely, if B is already in row echelon form, P and U are identities. The pattern is isomorphic to the iteration space, and B is the fitting matrix.

5   The Parametric Case

Parameters occurs mostly in loop bounds. They may also appear as strides and, more seldom, in the coefficients of subscript functions.

In the Array-OL formalism, the repetition loops must be square. Hence, their bound may be extracted diretcly from the program text. The extraction of the paving matrix is a simple derivative computation, which is an easy task for a competent computer algebra system.

Similarly, the B^k matrices are the result of a derivation, and may contain parameters.

There are no restrictions on the inner loops. For the construction of the pattern, one needs to know the vertices of the inner iteration domain. There are three cases:

- The bounds are constant: they can be extracted even if parametric.

- The bounds are affine expressions in other loop counters and parameters: the vertices can be computed with the help of the polylib.

- In other cases, there is no way of computing vertices, but the user may supply a bounding box.

The computation of the row echelon form can be done only if the matrix is known numerically, except in two cases: the matrix is 1×1 (it is its own normal form) or 2×2.

The row echelon form of $\begin{pmatrix} a & b \\ c & d \end{pmatrix}$ is $\begin{pmatrix} \gcd(a,b) & 0 \\ cu + dv & |(ad - bc)|/\gcd(a,b) \end{pmatrix}$ where u et v are the integers such that $au + bv = \gcd(a,b)$ whose existence is guaranteed by Bezout identity.

If none of these circumstance applies, the solution of last resort is to use one of the approximate schemes above. For instance, if the vertices of the inner iteration domain are available, it is possible, whatever the B matrix, to compute the vertices of the footprints and to enclose them in a rectangular box. The paving matrix is then the identity.

6 Extensions

The Syntol tool computes dependences; it is thus possible to check that the repetition loops are actually parallel. One must take care that Syntol will find dependences if temporary scalars are used in the code of the elementary transforms. These scalars must be expanded or privatized at code generation time.

Overlap between patterns (or, rather, between footprints) is another concern. For input arrays, overlap is just a cause of inefficiency, since some arrays cells will be copied several times to processors. Overlap for output arrays are more dangerous since they may induce non-determinism. The existence of overlap may be tested provided one stays inside the polytope model (affine loop bounds and indexing functions, with numerical coefficients and linear parameters). In the same context, it is possible to quantify the overhead by comparing the size of the pattern and the size of the real footprint using the `barvinok` library [4].

A   Computing the row echelon form of a matrix

For more details, see [3]. Let B be an arbitrary matrix of size $p \times q$.

1. At any stage of the computation, we have constructed two unitary matrices P and U such that:
$$B = PB'U, \ B' = \begin{bmatrix} H & 0 \\ C & D \end{bmatrix}$$

   where H is lower triangular with positive diagonal coefficients. Initially, P and U are identity matrices, H and C are empty and $D = B$. Let i be the index of the first row of C and D.

2. If D is null, the process stops.

3. If not, let j be the index of some non zero row of D. Let π_{ij} be the unitary matrix that permutes rows i and j of B'. Since π_{ij} is its own inverse, one can write:

$$B = (P\pi_{ij})(\pi_{ij}B')U,$$

and the new D has a non zero first row.

4. Let k be the index of a negative element in the first row of D. Let σ_k be the unit matrix with the k-th diagonal element set to -1. Since σ_k is its own inverse, one can write:

$$B = P(B'\sigma_k)(\sigma_k U),$$

and element k in the first row of D is now positive.

5. If all elements in the first row of D are positive, let l be the index of the smallest element, and let π_{il} be the matrix that interchange columns i and l of B'. Again:

$$B = P(B'\pi_{il})(\pi_{il}U)$$

and now the first element of the first row of D is smallest.

6. Let $m > i$ be the index of some nonzero element in the first row of D. Set $\alpha = B'_{im} \div B'_{ii}$. By construction, $\alpha > 0$. Let $\kappa_{im}(\alpha)$ be the identity matrix with $-\alpha$ added in position (i, m). It is easy to see that the inverse of $\kappa_{im}(\alpha)$ is $\kappa_{im}(-\alpha)$. Hence:

$$B = P(B'\kappa_{im}(\alpha))(\kappa_{im}(-\alpha)U)$$

and element B'_{im} has been replaced by $B'_{im} \bmod B'_{ii}$.

7. If the only non-zero element of the first row of D is the first element, then i can be increased by 1.

These transformations must be applied until no further progress is possible (i.e. when in case 2). Matrix B' is in the required form, and since all the elementary matrices π, σ and κ are unitary, the resulting P and U are unitary. In fact, P is even a permutation matrix.

References

[1] Pierre Boulet. Array-ol revisited, multidemensional intensive signal processing specification. Technical Report 6113, INRIA, January 2007.

[2] Eric Lenormand, 2007. Private communication.

[3] A. Schrijver. *Theory of linear and integer programming.* Wiley, NewYork, 1986.

[4] Rachid Seghir. *Méthodes de dénombrement de points entiers de polyèdres et applications à l'optimisation de programmes.* PhD thesis, ULP, December 2006.